1.

- a. (i) Generate four sets, each one consisting of 100 two-dimensional vectors, from the normal distributions with mean values  $[-10, -10]^T$ ,  $[-10, 10]^T$ ,  $[10, -10]^T$ ,  $[10, 10]^T$  and covariance matrices equal to 0.2\*I. These sets constitute the data set for a four-class two-dimensional classification problem (each set corresponds to a class).
- **a.** (ii) Compute the *Sw*, *Sb*, and *Sm* scatter matrices.
- **a.** (iii) Compute the value for the criterion *J*3.
- **b.** Repeat (a) when the mean vectors of the normal distributions that generate the data are  $[-1, -1]^T$ ,  $[-1, 1]^T$ ,  $[1, -1]^T$ ,  $[1, 1]^T$ .
- **c.** Repeat (a) when the covariance matrices of the normal distributions that generate the data are equal to 3 \* I.
- 2. The Fisher's discriminant ratio (FDR) is defined by:

$$FDR = \frac{(\mu_1 - \mu_2)^2}{\sigma_1^2 + \sigma_2^2}.$$

FDR is sometimes used to quantify the separability capabilities of **individual** features.

- **a.** (i) Generate two sets, each one consisting of 100 two-dimensional vectors, from the normal distributions with mean values  $[2, 4]^T$  and  $[2.5, 10]^T$  and covariance matrices equal to the  $2 \times 2$  identity matrix I. Their composition forms the data set for a two class two dimensional classification problem (each set corresponds to a class).
- a. (ii) Compute the value of the FDR index for both features.
- **b.** Repeat (a) when the covariance matrices of the normal distributions that generate the data are both equal to 0.25 \* I.
- **c.** Discuss the results.

- **a.** Generate an  $l \times N$  dimensional matrix X (l = 2 and N = 1000), whose columns are two-dimensional points lying around the line h:  $x_1 + x_2 = 0$ 
  - (i.e.,  $w = [1, 1]^T$  and  $w_0 = 0$ ), using the *generate\_hyper* function with parameters a = 10, e = 1 and sed = 0.
- **b.** Compute the principal components of the covariance of X as well as the corresponding variances (eigenvalues). Compare the direction of the first principal component with the direction vector of b (which is perpendicular to w) and draw your conclusions.

```
function X=generate_hyper(w,w0,a,e,N,sed)
l=length(w);

t=(rand(l-1,N)-.5)*2*a;
t_last=-(w(1:l-1)/w(l))'*t+2*e*(rand(1,N)-.5)-(w0/w(l));
X=[t; t_last];
%Plots for the 2d and 3d case
if(l==2)
    figure(1), plot(X(1,:),X(2,:),'.b')
elseif(l==3)
    figure(1), plot3(X(1,:),X(2,:),X(3,:),'.b')
end
figure(1), axis equal
```

MATLAB function named *generate\_hyper* that generates randomly l-dimensional points  $\mathbf{x}_i = [x_1(i), x_2(i), \dots, x_l(i)]^T$  around an (l-1)-dimensional hyperplane  $H \colon \mathbf{w}^T \mathbf{x} + w_0 = 0$ , where  $\mathbf{w} = [w_1, w_2, \dots, w_l]^T$ . More specifically, the function takes as inputs: (a) the parameter (column) vector  $\mathbf{w}$  for H ( $w_l \neq 0$ ), (b) the offset  $w_0$  for H, (c) a positive parameter a that defines the range [-a, a], where each one of the first (l-1) coordinates of the points is uniformly distributed, (d) the positive parameter e that defines the range [-e, e] of a uniformly distributed noise source, which is added to the term  $(-w_0 - \sum_{i=0}^{l-1} w_i x_i)/w_l$  to produce the lth coordinate, (e) the number N of points to be generated, and (f) the seed sed for the rand MATLAB function. It returns an  $l \times N$  dimensional matrix, X, whose columns contain the generated data points. In addition, the function plots the data points for l = 2, 3.